

# Complexity notes

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How long does this function execute?

```
def sum(list):  
    sum = 0  
    for elem in list:  
        sum += elem  
    return sum
```

(this function computes  $\sum a_i$  for  $\{a_1, \dots, a_n\}$ )

How long does this function execute?

```
def sum(list):  
    sum = 0  
    for elem in list:  
        sum += elem  
    return sum
```

Count the number of code lines:

$2n + 2$  if the length of list is  $n$

# Complexity of algorithms

Estimate of execution time  $T(n) = 2n+2$  for input  $n$

The big-O notation (asymptotic complexity):

Upper bound of  $T(n)$  is  $O(f(n))$

if  $T(n) \leq kf(n)$  for some  $k$  and  $n > n_0$

In this case,  $O(n)$  because we can choose (for example)  $k=3$  and  $n_0=1$ . Then  $2n+2 \leq 3n$  for  $n > 1$ .

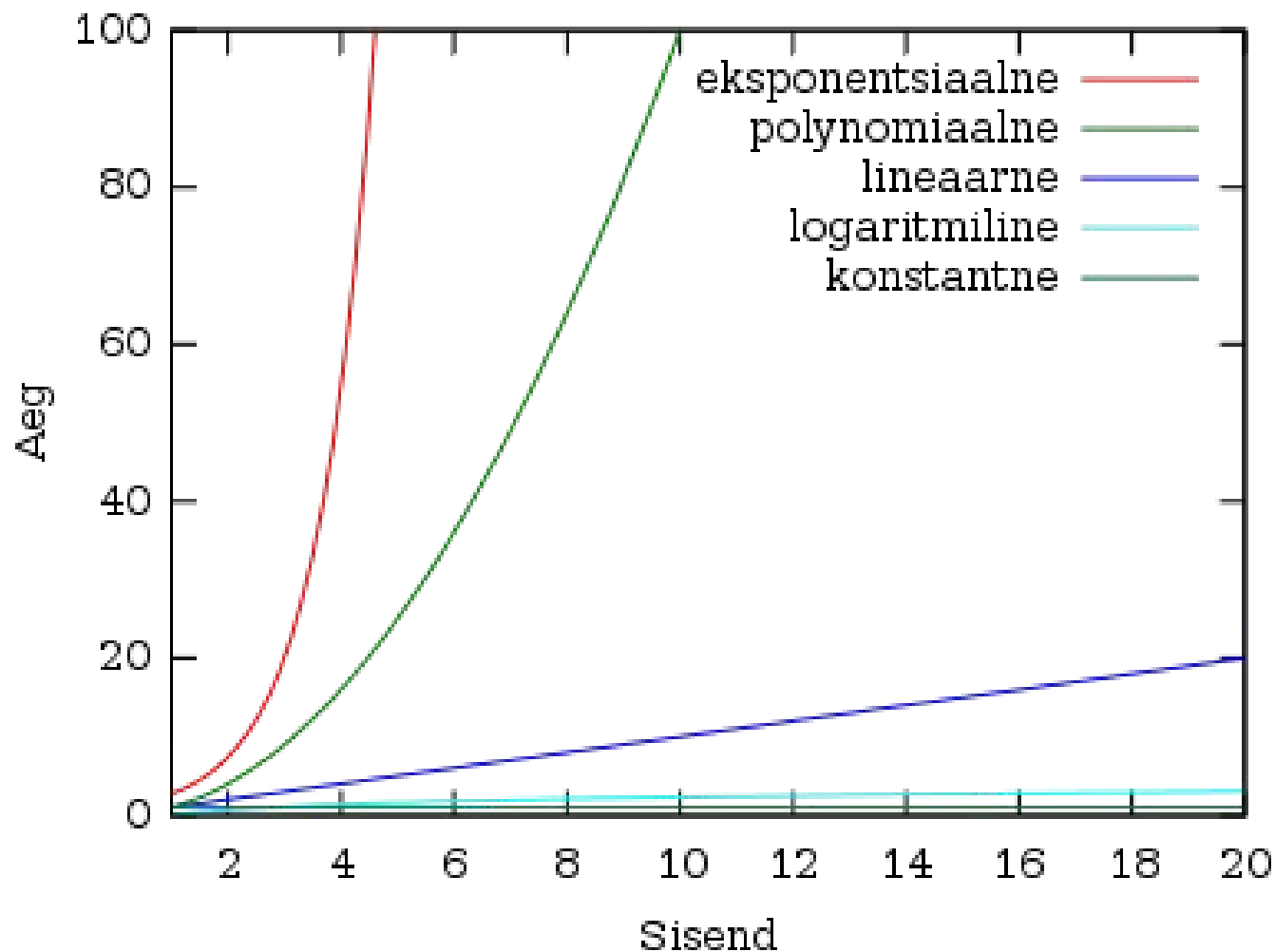
# Complexity of algorithms

The practical usefulness of the big-O notation:

	problem size increase	time increase
$O(n)$	5x	5x
$O(n^2)$	5x	25x
$O(2^n)$	5x	A LOT*

\*- for example, about  $10^{12}$  times if increasing from 10 to 50

# Common complexity classes



# Polynomial time

$a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$  is n-th order polynomial

These are also polynomials:

$2x$ ,  $6$ ,  $x^3$

$O(n^3)$  - polynomial complexity

the algorithm takes input size cubed time to run.

# Common simplifications

$k$  - constant

$$O(kn) = O(n)$$

$O(g(n)) = O(n^k)$  if  $g(n)$  is  $k$ -th order polynomial



# Complexity of problems

There are two main classes: P and NP

P - solvable in polynomial time

NP - ALSO solvable in polynomial time

(on a Non-deterministic Turing machine)

NP-complete - hardest problems in NP

# Complexity of problems

Practical implications:

- No polynomial time algorithms for NP-complete problems.
- A new problem is also NP-complete if we can easily convert a known NP-complete problem to the new problem.